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Information Spreading on Almost Torus Networks

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Abstract

Epidemic modeling has been extensively used in the last years in the field of telecommunications and computer networks. We consider the popular Susceptible-Infected-Susceptible spreading model as the metric for information spreading. In this work, we analyze information spreading on a particular class of networks denoted almost torus networks and over the lattice which can be considered as the limit when the torus length goes to infinity. Almost torus networks consist on the torus network topology where some nodes or edges have been removed. We find explicit expressions for the characteristic polynomial of these graphs and tight lower bounds for its computation. These expressions allow us to estimate their spectral radius and thus how the information spreads on these networks.

1 Introduction

There exists an abundant literature on epidemic modeling and in particular on epidemics on networks (see e.g., the books [1, 2, 3] and references therein). In the last decades, epidemic modeling has been extensively used in the field of telecommunications and computer networks. For example, epidemic models have been applied in order to analyze the spread of computer

viruses and worms [4], they have also been applied on epidemic routing for delay tolerant networks [5], etc. The structure of networks plays a critical role in the spread of a viral message and the authors of [6, 7, 8] have identified some conditions for successful performance of viral marketing. In particular, the authors of [7] have shown that the spectral radius of the graph determines the epidemic lifetime and its coverage. More recently on [9], the authors have studied the intertwined propagation of two competing “memes” (or viruses, rumors, etc.) in a composite network (individual agents are represented across two planes e.g. Facebook and Twitter). The authors have realized that the meme persistence δ and the meme strength β on each plane, and the spectral radius of the graph, completely determine which of the two competing “memes” prevail.

The topologies of focus in this work, torus network topologies, are commonly used in the production of high-end computing systems [10]. A number of supercomputers on the TOP500 list use three-dimensional torus networks [11]. For instance, IBM’s Blue Gene L [12, 13] and Blue Gene P [14]; Cray’s XT and XT3 [15] systems use three-dimensional torus networks for node communication. IBM Blue Gene/Q uses a five-dimensional torus network [16]. Fujitsu’s K computer and the PRIMEHPC FX10 use a proprietary six-dimensional torus interconnect called Tofu [17]. Torus networks are used because of a combination of their linear per node cost scaling and their competitive overall performance.

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In this work, we analyze information spreading on a particular class of networks denoted almost torus networks where we assume the popular Susceptible-Infected-Susceptible model as model of information spreading. Almost torus networks consist on the torus network topology where some nodes or edges have been removed. This situation can model the failure of some computer nodes or connections between the computer nodes. As we will see, in those graphs, the study of the spectral radius of the graph is determinant in order to analyze the information spreading. We find explicit expressions for the characteristic polynomial of these graphs and very tight lower bounds for its computation. These expressions allow us to estimate their spectral radius and thus how the information spreads on these networks.

The outline of the work is as follows. In Section 2, we recall some preliminary notions of graph theory. In Section 3, we present the Susceptible-Infected-Susceptible model of information spreading. Then, in Section 4, we analyze the information spreading in the almost torus network where one node has been removed and in Section 5 we extend our results to the cases when a set of nodes has been removed and when an edge has been removed from the torus network. In Section 6, we provide lower bounds for the two-dimensional torus network. In Section 7, we present numerical results that validate our analysis. Finally, in Section 8 we conclude.

2 Preliminaries

Let $\mathcal{G} = (\mathcal{V}, \mathcal{E})$ denote an undirected graph with no self-loops. We denote by $\mathcal{V} = \mathcal{V}(\mathcal{G}) = \{v_1, \dots, v_n\}$ the set of nodes and by $\mathcal{E} = \mathcal{E}(\mathcal{G}) \subseteq \mathcal{V} \times \mathcal{V}$ the set of undirected edges of \mathcal{G} . If $\{v_i, v_j\} \in \mathcal{E}(\mathcal{G})$ we call nodes v_i and v_j *adjacent* (or *neighbors*), which we denote by $v_i \sim v_j$. We define the set of neighbors of node v as $\mathcal{N}_v = \{w \in \mathcal{V} : \{v, w\} \in \mathcal{E}\}$. The *degree* of a node v , denoted by \deg_v , corresponds to the number of neighbors of v , i.e. the cardinality of the set \mathcal{N}_v . We define a *walk* of length k from v_0 to v_k to be an ordered sequence of nodes (v_0, v_1, \dots, v_k) such that $v_i \sim v_{i+1}$ for $i = 0, 1, \dots, k-1$. If $v_0 = v_k$, then the walk is closed.

Graphs can be algebraically represented via matrices. The *adjacency matrix* of an undirected graph \mathcal{G} , denoted by $A = A(\mathcal{G})$, is an $n \times n$ symmetric matrix defined entry-wise as

$$A_{ij} = \begin{cases} 1 & \text{if } v_i \text{ and } v_j \text{ are adjacent,} \\ 0 & \text{otherwise.} \end{cases}$$

We recall the well-known result that for $k \in \mathbb{N}$, A_{ij}^k is the number of paths of length k connecting the i -th and j -th vertices (proof by induction). Since A^0 is the identity matrix, we thus accept the existence of walks of length zero. We use I to denote the identity matrix, where its order is determined by the context.

We define the *Laplacian matrix* L for graphs without loops or multiple edges, as follows:

$$L_{ij} = \begin{cases} \deg_{v_i} & \text{if } v_i = v_j, \\ -1 & \text{if } v_i \text{ and } v_j \text{ are adjacent,} \\ 0 & \text{otherwise.} \end{cases}$$

We notice that the Laplacian of a graph can be written as $L = D - A$ where D is a diagonal matrix whose diagonal entries correspond to the degree of each node and A is the adjacency matrix.

The *spectral radius* of a graph \mathcal{G} , denoted $\rho(A)$, is the size of the largest eigenvalue (in absolute value) of the adjacency matrix of the graph, i.e. $\rho(A) = \max_i(|\lambda_i|)$. Since A is a symmetric matrix with non-negative entries, all its eigenvalues are real. The *characteristic polynomial* of \mathcal{G} , denoted $\phi(\mathcal{G}, x)$ is defined as $\det(xI - A)$, that corresponds to the characteristic polynomial of the adjacency matrix A . The *walk generating function*¹ $W(\mathcal{G}, x)$ is defined to be $(I - xA)^{-1}$. The ij -entry of $W(\mathcal{G}, x)$ will be written as $W_{ij}(\mathcal{G}, x)$. If \mathcal{S} is a subset of $\mathcal{V}(\mathcal{G})$ then $\mathcal{G} \setminus \mathcal{S}$ is the subgraph of \mathcal{G} induced by the vertices not in \mathcal{S} . We normally write $\mathcal{G} \setminus i$ instead of $\mathcal{G} \setminus \{i\}$ and $\mathcal{G} \setminus ij$ instead of $\mathcal{G} \setminus \{i, j\}$.

¹This can be viewed indifferently as a matrix with rational functions as entries, or as a formal power series in x over the ring of all polynomials in the matrix A .

3 Model of Information Spreading

We use the popular Susceptible-Infected-Susceptible (SIS) model of viral spreading [18] as the metric for information spreading. We remark that our results can be easily extended to the Susceptible-Infected₁-Infected₂-Susceptible (SI₁I₂S) model [9] of spreading over composite networks. We consider that each node can be in two possible states: susceptible (of being infected) or infected. We denote these two states as \mathcal{S} and \mathcal{I} , respectively.

The result presented on this section was first found in [19] and [20], through mean-field approximations of the Markov chain evolution of the 2^n possible states. We believe this alternative proof to be simpler and we presented here for completeness.

Consider a population of n nodes interconnected via an undirected graph $\mathcal{G} = (\mathcal{V}, \mathcal{E})$. Time is slotted. In each time slot, infected nodes attempt to contaminate their susceptible neighbors, where each infection attempt is successful with a probability β , independent of other infection attempts. The parameter β is called the virus birth rate (or meme strength as denoted in [9]). Each infected node recover in time slot t with probability δ . The parameter δ is called virus curing rate (or meme persistence). We notice that this parameter captures the meme persistence in an inverse way, i.e. a high δ means low persistence. This means a very contagious virus will be modeled with a low δ value.

We define $p_{i,t}$ as the probability that node i is infected at time t . We define $\zeta_{i,t+1}$ as the probability that node i will not receive infections from its neighbors in the next time-slot, which is given by

$$\zeta_{i,t+1} = \prod_{j \in \mathcal{N}_i} (1 - \beta p_{j,t}), \quad (1)$$

where \mathcal{N}_i denotes the set of neighbors of node i . This expression can be interpreted as the probability that none of the infection attempts is successful.

The probability for a node i of not being infected at time $t + 1$ depends on the conditioning event if the node i was or was not infected at time t :

1. For the first case, if the node i was infected at time t , then the probability for not being infected at time $t + 1$ is equal to $\delta \zeta_{i,t+1}$,
2. For the second case, if the node i was not infected at time t , then the probability for not being infected at time $t + 1$ is equal to $\zeta_{i,t+1}$.

From 1) and 2), the probability for a node i of not being infected at time $t + 1$ is equal to

$$1 - p_{i,t+1} = \zeta_{i,t+1}(1 - p_{i,t}) + \delta \zeta_{i,t+1} p_{i,t}.$$

Replacing from eq. (1), we obtain that

$$1 - p_{i,t+1} = (1 - p_{i,t}) \prod_{j \in \mathcal{N}_i} (1 - \beta p_{j,t}) + \delta p_{i,t} \prod_{j \in \mathcal{N}_i} (1 - \beta p_{j,t}). \quad (2)$$

We focus on the criterion based on the asymptotic stability of the disease-free equilibrium $p_i^*(t) = 0$ for all i . For doing this, we will make use of the following theorem.

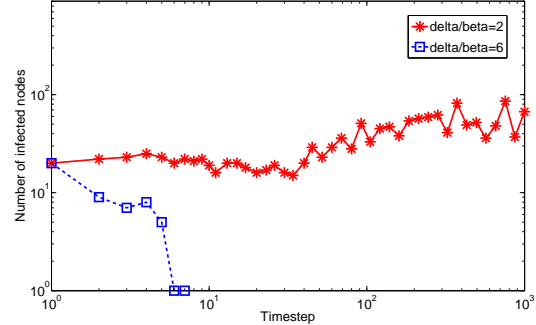


Figure 1: Number of infected nodes vs time.

Theorem 1 (Hirsch and Smale, 1974 [21]). *The system given by $\mathbf{p}_{t+1} = g(\mathbf{p}_t)$ is asymptotically stable at an equilibrium point \mathbf{p}^* , if the eigenvalues of the Jacobian $J = \nabla g(\mathbf{p})$ are less than 1 in absolute value, where*

$$J_{k,l} = [\nabla g(\mathbf{p}^*)]_{k,l} = \left. \frac{\partial p_{k,t+1}}{\partial p_{l,t}} \right|_{\mathbf{p}_t = \mathbf{p}^*}.$$

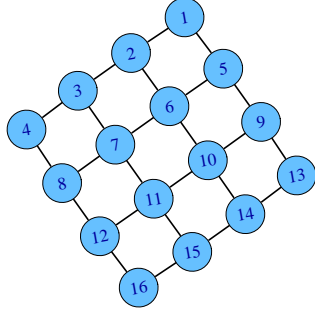


Figure 2: A two-dimensional grid of length 4.

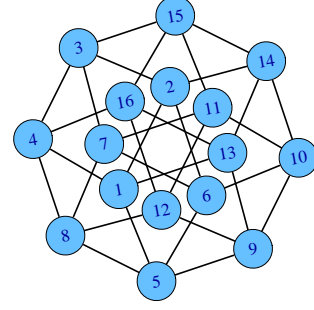


Figure 3: A two-dimensional torus of length 4.

We rewrite eq. (2) as follows:

$$p_{i,t+1} = 1 - (1 - p_{i,t}) \prod_{j \in \mathcal{N}_i} (1 - \beta p_{j,t}) - \delta p_{i,t} \prod_{j \in \mathcal{N}_i} (1 - \beta p_{j,t}).$$

Then

$$J_{k,l} = \begin{cases} 1 - \delta & \text{if } k = l, \\ \beta & \text{if } l \text{ is neighbor of } k, \\ 0 & \text{otherwise,} \end{cases} \quad (3)$$

where we recall that the evaluating point is $\mathbf{p}^* = 0$.

Eq. (3) can be written in a more compact way as $J = (1 - \delta)I + \beta A$. Using Theorem 1, we obtain that for an asymptotic stability of the disease-free equilibrium we need to impose that the eigenvalues of $(1 - \delta)I + \beta A$ are in absolute value smaller than 1, or equivalently,

$$\rho(A) < \frac{\delta}{\beta}. \quad (4)$$

In Fig. 1, we present two scenarios where we consider a total population of 900 nodes and an initial seeding of 20 nodes with the same probability of infection $\beta_1 = \beta_2 = 0.1$ but with different probabilities of recovery, $\delta_1 = 0.2$ for the solid curve and $\delta_2 = 0.6$ for the dashed curve. We consider a graph with spectral radius $\rho(A) = 4$. In the second scenario, as predicted by the analysis and equation (4), the virus or

information spreading dies out, however for the first scenario the virus or information spreading may continue as time increases.

4 Information Spreading over the Torus with one removed node

In a regular grid topology, each node in the network is connected with at most two neighbors along one or more directions (see e.g. Fig. 2). If the network is one-dimensional and we connect the first and last nodes, then the resulting topology consists on a chain of nodes connected in a circular loop, which is known as ring. In general, when an n -dimensional grid network is connected circularly in more than one dimension, the resulting network topology is a torus (see e.g. Fig. 3). In this work, we consider torus networks with the same number of nodes in every direction. From Fig. 2, by connecting each first node to the last node in each direction we obtain Fig. 3. For example, if we connect node 1 with node 4 in the horizontal direction and node 1 with node 13 in the vertical direction, we obtain the neighbors of node 1 on the torus (nodes 2, 4, 5, 13) as shown in Fig. 3.

From the previous section, we obtained that the spectral radius is an important quantity to study if our interest is the spreading of information (or virus spreading) through the network. The following proposition give us a relationship between the spectral radius and the degrees of the nodes.

Lemma 1. [22] *Let \deg_{\min} denote the minimum degree of \mathcal{G} , let $\overline{\deg}$ be the average degree of \mathcal{G} , and let \deg_{\max} be the maximum degree of \mathcal{G} . For every graph \mathcal{G} ,*

$$\max\{\overline{\deg}, \sqrt{\deg_{\max}}\} \leq \rho(A) \leq \deg_{\max}.$$

We observe that for a k -regular graph, its average degree and its maximum degree are equal to k , and thus k corresponds to its spectral radius. This means that for a d -dimensional torus, $2d$ corresponds to its spectral radius.

The previous lemma seems to prove that there is no interest in studying the spectral radius over the torus since it is a well-known quantity. However, if the torus network is modified (removing some of the nodes or edges to the graph) then the propagation of information will change (see Fig. 4 and Fig. 5). In the following analysis, we will give an explicit closed-form expression of these changes on the spreading.

First of all, we recall a well-known result in linear algebra. Cramer's rule [23] states that a system of n linear equations with n unknowns, represented in matrix multiplication form $Mx = y$, where the $n \times n$ matrix M has a nonzero determinant and the vector $x = (x_1, \dots, x_n)^T$ is the column vector of the unknown variables, has a unique solution, whose individual values for the unknowns are given by

$$x_i = \frac{\det(M_i)}{\det(M)}, \quad \forall i \in \{1, \dots, n\},$$

where M_i is the matrix formed by replacing the i -th column of M by the column vector y .

We use Cramer's rule in the next lemma, in order to establish the connection between the characteristic polynomials and walk generating functions. This connection will allow us to compute the spectral radius of the modified torus when we remove a node.

Lemma 2. [24] *For any graph \mathcal{G} we have*

$$x^{-1}W_{ii}(\mathcal{G}, x^{-1}) = \phi(\mathcal{G} \setminus i, x) / \phi(\mathcal{G}, x).$$

Proof. We have that $W(\mathcal{G}, x) = (I - xA)^{-1} = x^{-1}(x^{-1}I - A)^{-1}$. The entries of $(x^{-1}I - A)^{-1}$ are given by Cramer's rule (by noting that $(x^{-1}I - A)^{-1}$ corresponds to the matrix M such that $(x^{-1}I - A)M = I$). The i -th diagonal entry of $(x^{-1}I - A)^{-1}$ is given by the i -th principal diagonal minor of $(x^{-1}I - A)$ divided by $\det(x^{-1}I - A) = \phi(\mathcal{G}, x^{-1})$. We note that the i -th principal diagonal minor of $(x^{-1}I - A)$ is $\phi(\mathcal{G} \setminus i, x^{-1})$, and so the lemma follows immediately from this. \square

We are interested on finding the spectral radius of the almost regular torus, in order to analyze the information spreading on this topology. In the case of the removal of one node, this is equivalent to find $\phi(\mathcal{G} \setminus i, x)$. From Lemma 2, we need to know the characteristic polynomial of the regular torus $\phi(\mathcal{G}, x)$ and the diagonal entries of the walking generating function of the regular torus $W_{ii}(\mathcal{G}, x^{-1})$ since

$$\phi(\mathcal{G} \setminus i, x) = x^{-1}W_{ii}(\mathcal{G}, x^{-1})\phi(\mathcal{G}, x). \quad (5)$$

In the next proposition, we give an explicit expression for the characteristic polynomial of the two-dimensional torus network of length m .

Proposition 1. *The characteristic polynomial of the two-dimensional torus network, denoted \mathcal{T}_m^2 , is given by*

$$\phi(\mathcal{T}_m^2, x) = \prod_{1 \leq i, j \leq m} (x - 2 \cos(2\pi i/m) - 2 \cos(2\pi j/m)).$$

Proof. Consider R_m the ring graph which has edge set $\{(u, u+1) : 1 \leq u < m\} \cup \{(m, 1)\}$. The Laplacian of R_m has eigenvectors [25]:

$$x_k(u) = \sin(2\pi ku/m) \text{ and } y_k(u) = \cos(2\pi ku/m). \quad (6)$$

for $k \leq m/2$. Both of these eigenvectors have eigenvalue $\lambda_k = 2 - 2 \cos(2\pi k/m)$. We notice that x_0

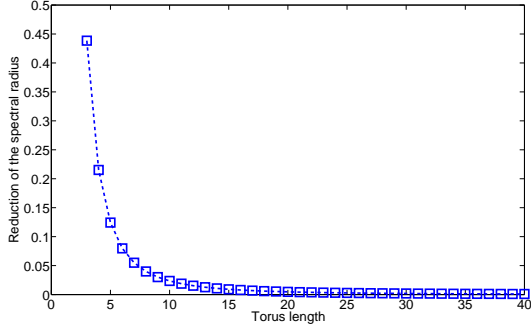


Figure 4: Spectral radius reduction by removing one node vs torus length.

should be ignored and y_0 is the all 1s vector. If m is even, then also $x_{m/2}$ should be ignored.

We recall that the product of two graphs $\mathcal{G} = (\mathcal{V}, \mathcal{E})$ and $\mathcal{H} = (\mathcal{W}, \mathcal{F})$, denoted by $\mathcal{G} \times \mathcal{H}$, corresponds to the graph with vertex set $\mathcal{V} \times \mathcal{W}$ and edge set $((v_1, w_1), (v_2, w_2))$ where $(v_1, v_2) \in \mathcal{E}$ and $(w_1, w_2) \in \mathcal{F}$. If \mathcal{G} has Laplacian eigenvalues $\lambda_1, \dots, \lambda_m$ and eigenvectors p_1, \dots, p_m ; and \mathcal{H} has Laplacian eigenvalues μ_1, \dots, μ_m and eigenvectors q_1, \dots, q_m ; then for each $1 \leq i \leq m$ and $1 \leq j \leq m$, $\mathcal{G} \times \mathcal{H}$ has an eigenvector z of eigenvalue $\lambda_i + \mu_j$ given by $z(v, w) = p_i(v)q_j(w)$.

In our case, the two-dimensional torus can be written as the product of two rings, $\mathcal{G} = \mathcal{H} = R_m$, and then $\lambda_i = \mu_i = 2 - 2\cos(2\pi i/m) \quad \forall 1 \leq i \leq m$. Thus the eigenvalues of the Laplacian of the two-dimensional torus network of length m are given by

$$\lambda_i + \mu_j = 4 - 2\cos(2\pi i/m) - 2\cos(2\pi j/m) \quad \forall 1 \leq i, j \leq m.$$

We recall that the eigenvalues of a matrix M are the solutions λ to the equation $\det(M - \lambda I) = 0$. For

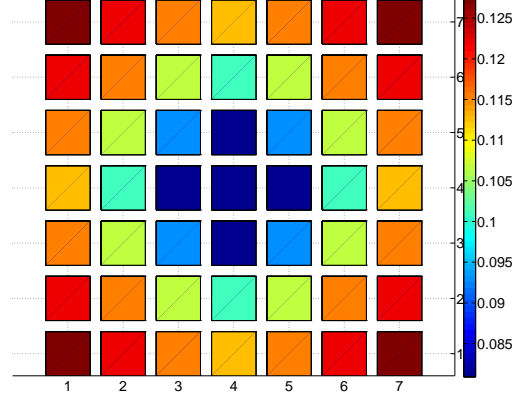


Figure 5: Spectral radius reduction by removing two nodes from a torus network of length 7. One of the removed nodes is the central node. We remove another node and we compute the spectral radius reduction and we assign the value of this reduction at the removed node's position.

a k -regular graph of n nodes we have

$$\begin{aligned} \det(L - \lambda I) &= \det(D - A - \lambda I) \\ &= \det(-[A - (k - \lambda)I]) = (-1)^n \det(A - (k - \lambda)I), \end{aligned}$$

which means that λ is an eigenvalue of L if and only if $k - \lambda$ is an eigenvalue of A .

In our case, the two-dimensional torus network is a 4-regular graph, and thus the eigenvalues of the two-dimensional torus network of length m are given by $2\cos(2\pi i/m) + 2\cos(2\pi j/m)$ for all $1 \leq i, j \leq m$. Thus we conclude that the characteristic polynomial is equal to

$$\phi(\mathcal{T}_m^2, x) = \prod_{1 \leq i, j \leq m} (x - 2\cos(2\pi i/m) - 2\cos(2\pi j/m)).$$

We notice from the proof that the eigenvectors of the Laplacian coincide with the eigenvectors of the adjacency matrix and are given by the Kronecker product of the eigenvectors given by (6). \square

In the following proposition, we find an explicit expression for the diagonal entries of the walk generating function.

Proposition 2. *The diagonal entries of the walk generating function of the two-dimensional torus of length m are:*

$$\begin{aligned}
W_{ii}(\mathcal{T}_m^2, x) &= \\
\sum_{\ell \geq 0} \frac{x^\ell}{m^2} \sum_{1 \leq i, j \leq m} 4^\ell (\cos(\pi(i+j)/m))^\ell (\cos(\pi(i-j)/m))^\ell &= \\
\sum_{\ell \geq 0} \frac{x^{-\ell}}{m^2} \sum_{1 \leq i, j \leq m} 4^\ell (\cos(\pi(i+j)/m))^\ell (\cos(\pi(i-j)/m))^\ell, & \\
\phi(\mathcal{T}_m^2, x) &= \prod_{1 \leq i, j \leq m} (x - 2 \cos(2\pi i/m) - 2 \cos(2\pi j/m)).
\end{aligned} \tag{7}$$

Proof. We recall from Section 2 that A_{ii}^ℓ corresponds to the closed walks of length ℓ . Since each node is indistinguishable on the torus network, we know that $\text{tr}(A^\ell) = n \times \#$ of closed walks of length ℓ , where $n = m^2$ is the total number of nodes. But we also have that

$$\begin{aligned}
\text{tr}(A^\ell) &= \sum_{k=1}^n (\lambda_i(A))^\ell, \\
&= \sum_{1 \leq i, j \leq m} (2 \cos(2\pi i/m) + 2 \cos(2\pi j/m))^\ell, \\
&= \sum_{1 \leq i, j \leq m} (4 \cos(\pi(i+j)/m) \cos(\pi(i-j)/m))^\ell.
\end{aligned}$$

Thus

$$A_{ii}^\ell = \frac{1}{m^2} \sum_{1 \leq i, j \leq m} (4 \cos(\pi(i+j)/m) \cos(\pi(i-j)/m))^\ell. \tag{8}$$

Let us notice that

$$W_{ii}(\mathcal{G}, x) = [(I - xA)^{-1}]_{ii} = \left[\sum_{n \geq 0} (xA)^n \right]_{ii} = \sum_{\ell \geq 0} x^\ell A_{ii}^\ell W_{ii}(\mathcal{T}_m^d, x^{-1}) = \tag{9}$$

From eq. (8) and eq. (9) we conclude eq. (7). \square

From Proposition 1 and Proposition 2, we obtain the following theorem.

Theorem 2. *The characteristic polynomial of the two-dimensional torus network of length m where one*

node has been removed is given by:

$\phi(\mathcal{T}_m^2 \setminus i, x) = x^{-1} W_{ii}(\mathcal{T}_m^2, x^{-1}) \phi(\mathcal{T}_m^2, x)$, where

$$W_{ii}(\mathcal{T}_m^2, x^{-1}) =$$

$$\sum_{\ell \geq 0} \frac{x^{-\ell}}{m^2} \sum_{1 \leq i, j \leq m} 4^\ell (\cos(\pi(i+j)/m))^\ell (\cos(\pi(i-j)/m))^\ell,$$

$$\phi(\mathcal{T}_m^2, x) = \prod_{1 \leq i, j \leq m} (x - 2 \cos(2\pi i/m) - 2 \cos(2\pi j/m)).$$

We observe that all the previous calculations do not depend on the particular removed node i . This means that the spreading of information over the modified torus, if we remove one node, is not affected by the position of the removed node. The importance is that one and only one node is removed. In the next section, we will see that this is very different from the case of the removal of two nodes.

The previous results can also be derived for the d -dimensional torus network. Since the proofs are similar to the proofs of Proposition 1 and Proposition 2 we do not include them here.

Theorem 3. *The characteristic polynomial of the d -dimensional torus network of length m where one node has been removed is given by:*

$$\phi(\mathcal{T}_m^d \setminus i, x) = x^{-1} W_{ii}(\mathcal{T}_m^d, x^{-1}) \phi(\mathcal{T}_m^d, x),$$

where the diagonal entries of the walk generating function are

$$\sum_{\ell \geq 0} \frac{x^{-\ell}}{m^d} \sum_{1 \leq i_1, i_2, \dots, i_d \leq m} (2 \cos(2\pi i_1/m) + \dots + 2 \cos(2\pi i_d/m))^\ell,$$

and the characteristic polynomial of the d -dimensional torus network is given by

$$\phi(\mathcal{T}_m^d, x) =$$

$$\prod_{1 \leq i_1, \dots, i_d \leq m} (x - 2 \cos(2\pi i_1/m) - \dots - 2 \cos(2\pi i_d/m)).$$

5 Information Spreading over the Torus with a set of removed nodes

Following a similar approach to Lemma 2, Godsil [24] is able to prove the following theorem.

Theorem 4. [Nodes removal][24] *Let \mathcal{S} be a subset of s nodes from the graph \mathcal{G} . Then*

$$x^{-s} \det W_{\mathcal{S}, \mathcal{S}}(\mathcal{G}, x^{-1}) = \phi(\mathcal{G} \setminus \mathcal{S}, x) / \phi(\mathcal{G}, x),$$

where $W_{\mathcal{S}, \mathcal{S}}(\mathcal{G}, x)$ denotes the submatrix of $W(\mathcal{G}, x)$ with rows and columns indexed by the elements of \mathcal{S} .

We observe that if \mathcal{S} consists of two nodes i and j then

$$\det W_{\mathcal{S}, \mathcal{S}}(\mathcal{G}, x) = W_{ii}(\mathcal{G}, x)W_{jj}(\mathcal{G}, x) - W_{ij}(\mathcal{G}, x)W_{ji}(\mathcal{G}, x). \quad (10)$$

and since \mathcal{G} is an undirected graph, $W_{ij}(\mathcal{G}, x) = W_{ji}(\mathcal{G}, x)$.

For the case of the removal of two nodes, from Theorem 4 and eq. (10), we obtain the following corollary.

Corollary 1. [24] *For any graph \mathcal{G} we have:*

$$x^{-1}W_{ij}(\mathcal{G}, x^{-1}) = \frac{\sqrt{\phi(\mathcal{G} \setminus i, x)\phi(\mathcal{G} \setminus j, x) - \phi(\mathcal{G}, x)\phi(\mathcal{G} \setminus ij, x)}}{\phi(\mathcal{G}, x)}. \quad (11)$$

The next corollary is extremely important since it guarantees that independently of the number of nodes we remove from the torus graph, we can restrict our study of the characteristic polynomials to the case of the removal of two nodes.

Corollary 2. [24] *If \mathcal{C} is a subset of $\mathcal{V}(\mathcal{G})$ then $\phi(\mathcal{G} \setminus \mathcal{C}, x)$ is determined by the polynomials $\phi(\mathcal{G} \setminus \mathcal{S}, x)$ where \mathcal{S} ranges over all subsets of \mathcal{C} with at most two vertices.*

From Corollary 1, we obtain that

$$\begin{aligned} \phi(\mathcal{G} \setminus ij, x) &= \\ \frac{\phi(\mathcal{G} \setminus i, x)\phi(\mathcal{G} \setminus j, x)}{\phi(\mathcal{G}, x)} - \phi(\mathcal{G}, x) (x^{-1}W_{ij}(\mathcal{G}, x^{-1}))^2. \end{aligned} \quad (12)$$

This implies that the only unknowns to compute the removal of a set of nodes over the torus, in particular for a set of two nodes, are given by W_{ij} , which can be represented as the number of walks between node i and node j . The following theorem provides us a way to compute the values W_{ij} .

Theorem 5. [26] *For any two vertices i and j in the graph \mathcal{G} and any non-negative integer ℓ , we have that the number of walks of length ℓ , denoted $W_{ij}(\mathcal{G}, \ell)$, is given by*

$$W_{ij}(\mathcal{G}, \ell) = \sum_{\theta} u_{\theta}(i)^T \theta^{\ell} u_{\theta}(j),$$

where the sum is over all eigenvalues θ of \mathcal{G} and $u_{\theta}(i)$ denotes the i -th row of U_{θ} where U_{θ} is the matrix whose columns form an orthonormal basis for the eigenspace belonging to θ .

From the previous theorem, we conclude that the walk generating function can be written as

$$W_{ij}(\mathcal{G}, x^{-1}) = \sum_{\ell \geq 0} x^{-\ell} \sum_{\theta} u_{\theta}(i)^T \theta^{\ell} u_{\theta}(j).$$

In order to compute the walk generating function of the two-dimensional torus, we recall that the ring of length m has eigenvectors:

$$x_k(u) = \sin(2\pi ku/m) \quad \text{and} \quad y_k(u) = \cos(2\pi ku/m).$$

for $k \leq m/2$. Both of these eigenvectors have eigenvalue $\lambda_k = 2 - 2\cos(2\pi k/m)$. Here x_0 should be ignored and y_0 is the all 1s vector. If m is even, then also $x_{m/2}$ should be ignored. We denote the matrix of eigenvectors of the ring of length m as V . Then the two-dimensional torus has matrix of eigenvectors $Z = V \otimes V$ where \otimes denotes the Kronecker product. Since the matrix is symmetric the eigenvectors form an orthogonal basis and we may normalize its eigenvectors to obtain an orthonormal basis. We call this orthonormal basis \hat{Z} . We consider the matrix of eigenvalues of the two-dimensional torus, denoted by Λ , which are given by

$$\lambda_i + \mu_j = 2\cos(2\pi i/m) + 2\cos(2\pi j/m) \quad \forall 1 \leq i, j \leq m.$$

From here we obtain the number of closed walks of the two-dimensional torus between node i and node j .

Following a similar approach to the previous case of the removal of one node, Godsil [24] is able to prove the following theorem which deals with the removal of one edge of the graph.

Theorem 6 (Edge removal). [24] *Let $e = \{i, j\}$ be an edge in \mathcal{G} . Then*

$$\begin{aligned} \phi(\mathcal{G}, x) &= \phi(\mathcal{G} \setminus e, x) - \phi(\mathcal{G} \setminus ij, x) \\ &\quad - 2\sqrt{\phi(\mathcal{G} \setminus i, x)\phi(\mathcal{G} \setminus j, x) - \phi(\mathcal{G}, x)\phi(\mathcal{G} \setminus ij, x)}. \end{aligned} \quad (13)$$

We observe that all the previous terms in eq. (13) except $\phi(\mathcal{G} \setminus e, x)$ are known. This implies that the characteristic polynomial of the two-dimensional torus with one edge removed is completely determined by the previous expressions for the torus with one node removed and two nodes removed.

6 Lower bounds for the two-dimensional torus networks

In this section, we will compare the previous expressions with the case of the infinite lattice. We notice that this scenario can be seen as the case where the length of the torus goes to infinity.

Proposition 3. *The diagonal entries of the walk generating function over the lattice \mathcal{L}^2 are given by*

$$W_{ii}(\mathcal{L}^2, x) = \sum_{\ell \geq 0} x^{2\ell} \binom{2\ell}{\ell}^2. \quad (14)$$

Proof. The number of closed walks of length 2ℓ over the two-dimensional lattice² \mathcal{L}^2 is equal to (see e.g. EIS A002894):

$$\sum_{i=0}^{\ell} \frac{(2\ell)!}{i!i!(\ell-i)!(\ell-i)!} = \binom{2\ell}{\ell}^2.$$

² We notice that to consider even length is not restrictive since every closed walk over the lattice network has to have even length.

In the following, we give a lower bound to the walk generating function of the two-dimensional torus with one removed node.

Proposition 4. *For the two-dimensional torus, we have*

$$x^{-1}W_{ii}(\mathcal{T}_m^2, x^{-1}) \geq \frac{1}{x} \frac{4\pi}{e^4} \sum_{\ell \geq 0} \frac{1}{\ell} \left(\frac{4}{x}\right)^{2\ell}.$$

Proof. The Stirling approximation formula tell us that

$$\sqrt{2\pi\ell} \left(\frac{\ell}{e}\right)^{\ell} \leq \ell! \leq e\sqrt{\ell} \left(\frac{\ell}{e}\right)^{\ell}.$$

Then applying it to the binomial coefficient we get the lower bound

$$\binom{2\ell}{\ell} = \frac{(2\ell)!}{\ell!\ell!} \geq \frac{\sqrt{2\pi 2\ell} \left(\frac{2\ell}{e}\right)^{2\ell}}{e\sqrt{\ell} \left(\frac{\ell}{e}\right)^{\ell} e\sqrt{\ell} \left(\frac{\ell}{e}\right)^{\ell}} = \frac{\sqrt{4\pi}}{e^2} \frac{1}{\sqrt{\ell}} 4^{\ell}. \quad (15)$$

We know that the number of walks over the torus network is greater than the number of walks over the lattice network since every walk over the lattice network can be mapped into a walk over the torus network. We notice that the contrary is not true in general since for example following $m+1$ steps in the same direction will be a closed walk over the torus of length m but not a closed walk over the lattice. Therefore, from eq. (14) and eq. (15), we have that

$$\begin{aligned} x^{-1}W_{ii}(\mathcal{T}_m^2, x^{-1}) &\geq x^{-1}W_{ii}(\mathcal{L}^2, x^{-1}) = \\ &= x^{-1} \sum_{\ell \geq 0} x^{-2\ell} \binom{2\ell}{\ell}^2 \geq \frac{1}{x} \frac{4\pi}{e^4} \sum_{\ell \geq 0} \frac{1}{\ell} \left(\frac{4}{x}\right)^{2\ell}. \end{aligned}$$

□

Proposition 4 allow us to obtain also a lower bound for the characteristic polynomial of the two-dimensional torus with one removed node.

The lower bound is given by

$$\phi(\mathcal{T}_m^2 \setminus i, x) \geq \phi(\mathcal{T}_m^2, x) \frac{1}{x} \frac{4\pi}{e^4} \sum_{\ell \geq 0} \frac{1}{\ell} \left(\frac{4}{x}\right)^{2\ell} \quad \text{where}$$

$$\phi(\mathcal{T}_m^2, x) = \prod_{1 \leq i, j \leq m} (x - 2 \cos(2\pi i/m) - 2 \cos(2\pi j/m)).$$

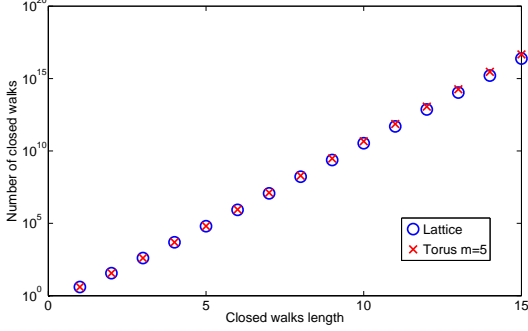


Figure 6: Number of closed walks vs closed walks length.

The case of the number of walks between any two nodes over the two-dimensional torus of length m is more involved. We define the following function which allow us to map each node on the two-dimensional torus of length m to a node on the lattice $\mathbb{Z} \times \mathbb{Z}$. The mapping function $h : \{1, \dots, m^2\} \rightarrow \mathbb{Z} \times \mathbb{Z}$ is as follows:

$$h(u) = (x_u, y_u) \quad \text{where} \\ x_u \equiv u - 1 \pmod{m} \quad \text{and} \quad y_u = \left\lfloor \frac{u}{m} \right\rfloor.$$

Proposition 5. *The number of walks of length ℓ over the two-dimensional lattice from the origin $(0,0)$ to (a,b) ($\ell + a + b$ is even) is equal to*

$$\sum_{i=b}^{b+(\ell-b)/2} \frac{\ell!}{i!(i-b)!\left(\frac{\ell+a+b}{2}-i\right)!\left(\frac{\ell+b-a}{2}-i\right)!}. \quad (16)$$

Proof. We consider n_U the number of steps you go up, n_D the number of steps you go down, n_L the number of steps you go left, n_R the number of steps you go right. The number of walks of length ℓ from $(0,0)$ to (a,b) is equal to the number of ways you can choose ℓ steps between up, down, left, and right, such that $n_U = n_D + b$, $n_R = n_L + a$, and $n_U + n_D + n_L + n_R = \ell$. If we fix $n_U = i$ then $n_D = i - b$, $n_L = \left(\frac{\ell+b-a}{2} - i\right)$, $n_R = \left(\frac{\ell+a+b}{2} - i\right)$ and the number

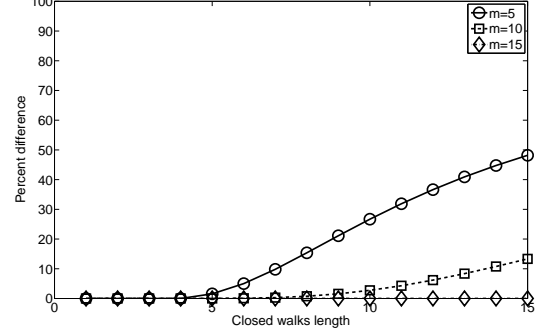


Figure 7: Percent difference on the number of closed walks between the torus and the lattice vs closed walks length.

of closed walks of length ℓ with i steps going up is equal to $\frac{\ell!}{i!(i-b)!\left(\frac{\ell+a+b}{2}-i\right)!\left(\frac{\ell+b-a}{2}-i\right)!}$. Summing up for all possible values of n_U , the number of paths is equal to eq. (16). \square

From Proposition 5 and the mapping function h , we obtain a lower bound on the number of walks of length ℓ on the two-dimensional torus \mathcal{T}_m^2 between two different nodes i and j by mapping i and j to $h(i) = (x_i, y_i)$ and $h(j) = (x_j, y_j)$ and computing the number of walks from $(0,0)$ to $(x_j - x_i, y_j - y_i)$. We observe that the previous lower bound is tight only when $\ell + a + b$ is even.

7 Numerical Simulations

In Fig. 6, we compute the number of closed walks over the torus of length $m = 5$ and over the lattice vs the length of the closed walks ℓ . We notice that for small values of ℓ to approximate the number of closed walks over the torus is more precise than for large values of ℓ . This fact is compensated since in the characteristic polynomial the important terms are the terms of smaller order. In Fig. 7, we compute the difference on percent error of this approximation vs the length of the closed walks ℓ .

In Fig. 8, we compute the number of closed walks over the torus and over the lattice by varying the

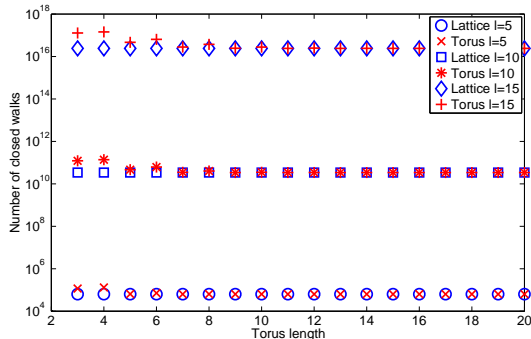


Figure 8: Number of closed walks vs torus length.

torus length m and keeping constant the length of the closed walks ℓ . We notice that for small values of the torus length m the approximation is not tight, however it becomes tighter relatively fast. In Fig. 9, we compute the difference on percent error of this approximation vs the torus length m .

8 Conclusions

In this work, we have analyzed information spreading on almost torus networks assuming the Susceptible-Infected-Susceptible spreading model as a metric of information spreading. Almost torus networks consist on the torus network topology where some nodes or edges have been removed. We have provided analytical expressions for the characteristic polynomial of these graphs and we have provided as well tight lower bounds for its computation. Using these expressions we are able to estimate their spectral radius and thus to know how the information spreads on these networks. Simulations results that validated our analysis are presented.

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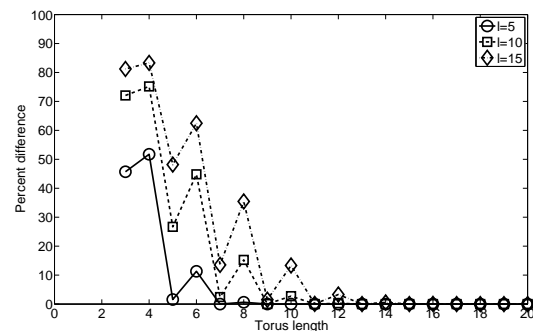


Figure 9: Percent difference on the number of closed walks between the torus and the lattice vs torus length.

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